#### Redes Generativas Adversárias

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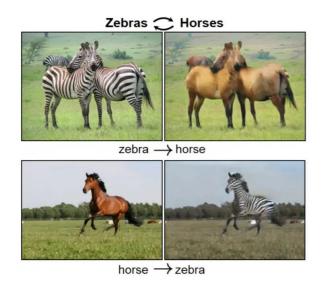
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# **Tópicos**

- DCGAN
- PIX2PIX
- Prática

## **Deep Fakes**

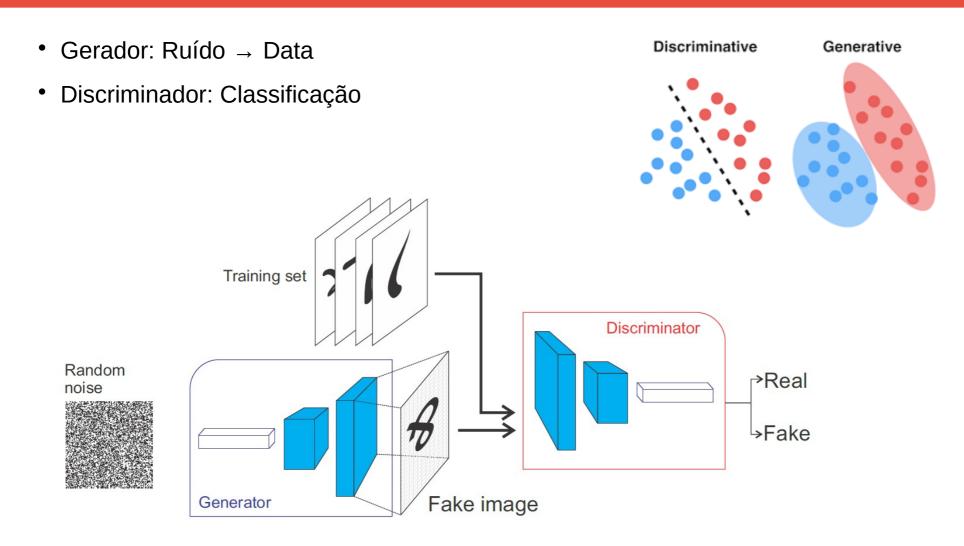
- Generalização: Dados Sintéticos gerados a partir do aprendizado da distribuição real do dado
- Aplicações
  - Filmes (Cenários Sintéticos)
  - Fotografia (Estimação de Pose, Coloração Artificial, Redução de Ruído)
  - Troca de Contexto (Zebra->Cavalo)
  - •





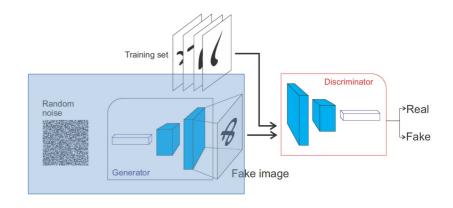


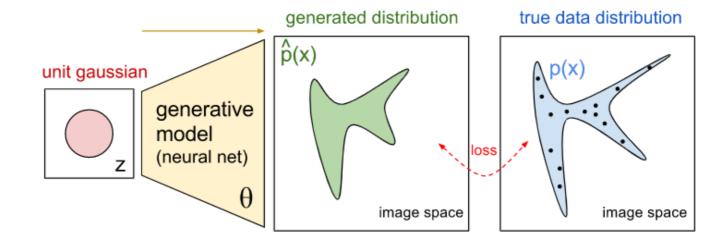
# Generative Adversarial Networks (GAN's)



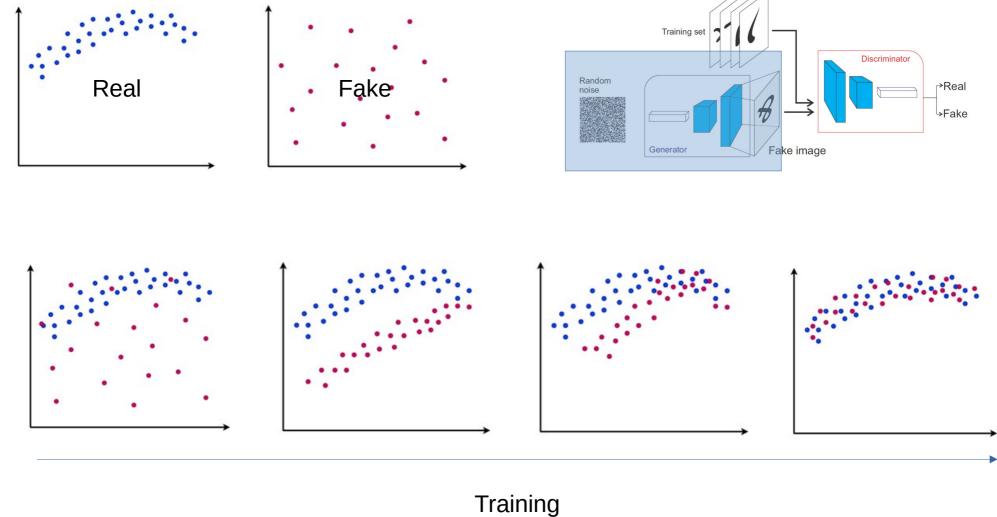
#### **Modelo Generativo**

Aprende a distribuição do dado



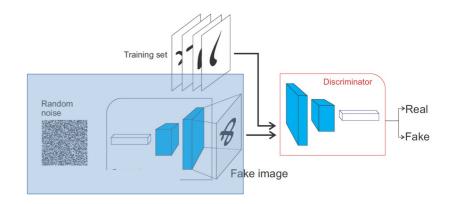


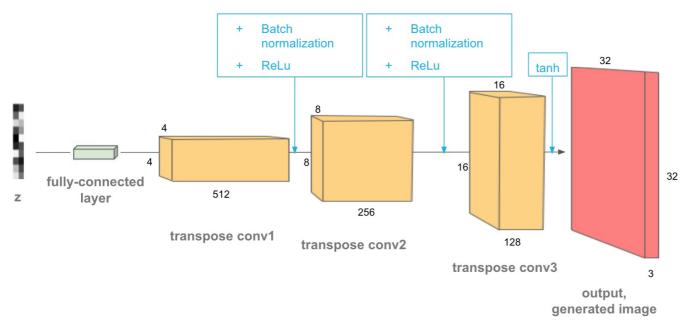
### **Modelo Generativo**



#### **Modelo Generativo Profundo**

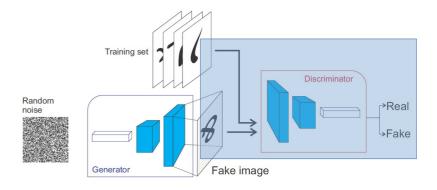
- Camadas Des-Convolucionais (upsampling)
  - Ruído → Imagens Sintéticas

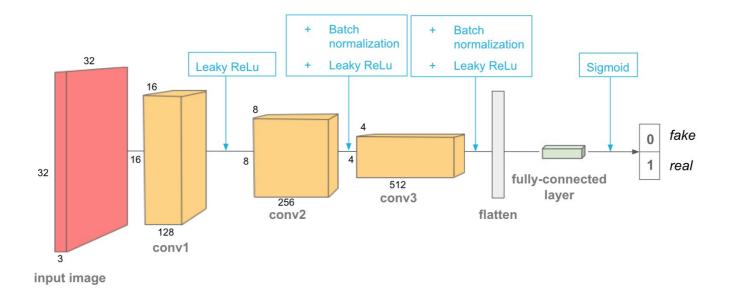




#### **Modelo Discriminante**

- Classificação: Falso ou Real ?
  - CNN

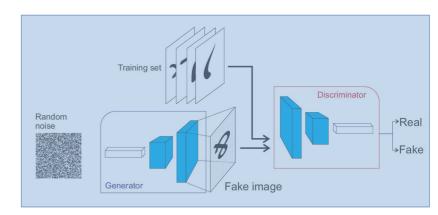


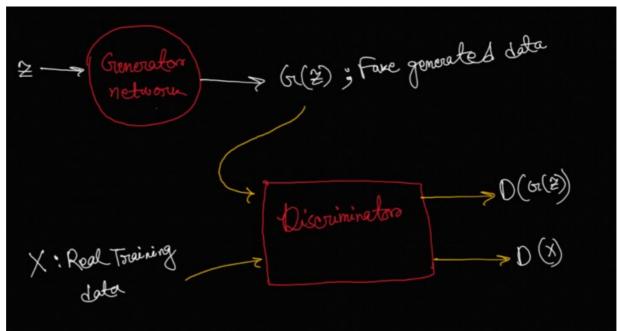


#### **Treinamento Adversário**

• Erro Adversário (Min-Max)

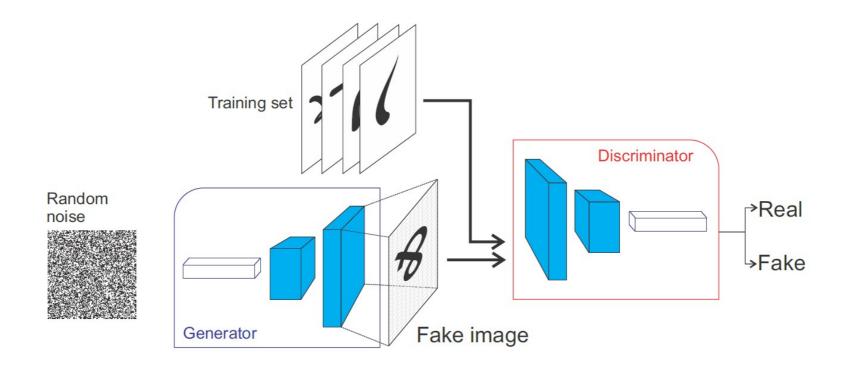
$$\min_{G} \max_{D} V(D,G) = \mathbb{E}_{x \sim p_{data}}[\log D(x)] + \mathbb{E}_{z \sim p_z(z)}[\log(1 - D(G(z)))]$$





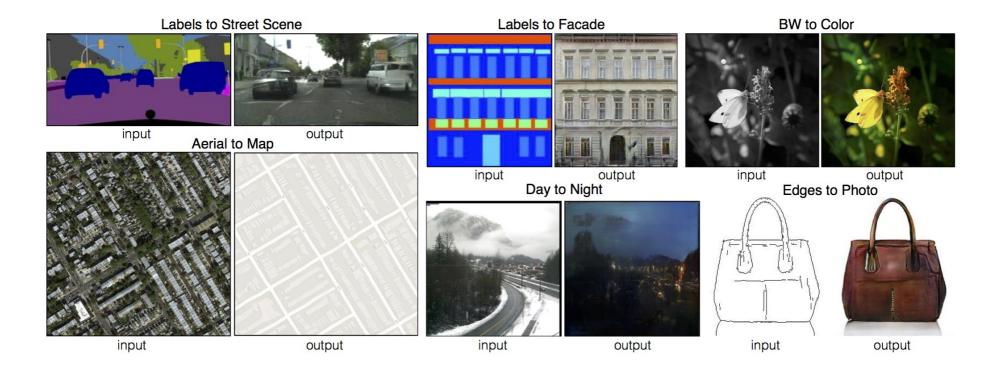
#### Let's Code

• [LINK]



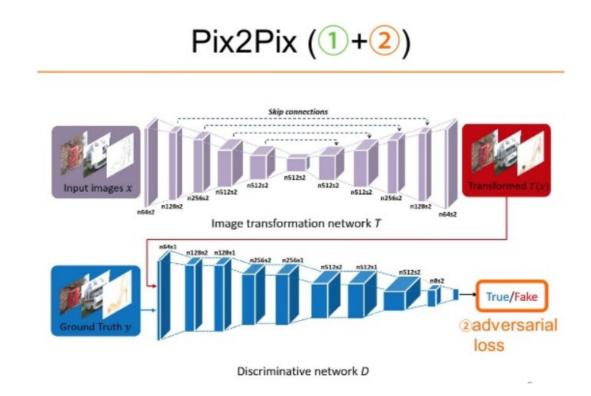
#### Pix2Pix

• Transformação de Contexto (Image Translation)



#### Pix2Pix

- Modelo Generativo: Arquitetura Encoder-Decoder (i.e U-Net)
- Base de Dados Pareada (Origem->Destino)





#### Let's Code

LINK: Lecture\_12-pix2pix.ipynb

